

Estimating large scale dynamic mountain glacier states with numerical modelling and data assimilation

Introducing COMBINE

Patrick Schmitt¹, Fabien Maussion¹, Philipp Gregor², Daniel Goldberg³

¹Department of Atmospheric and Cryospheric Sciences, University of Innsbruck, Innsbruck, Austria

²Meteorologisches Institut, Ludwig-Maximilians-Universität München, München, Germany

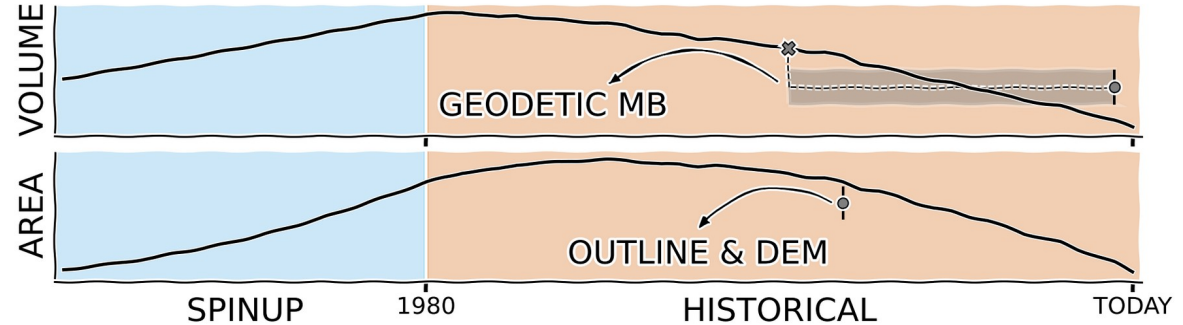
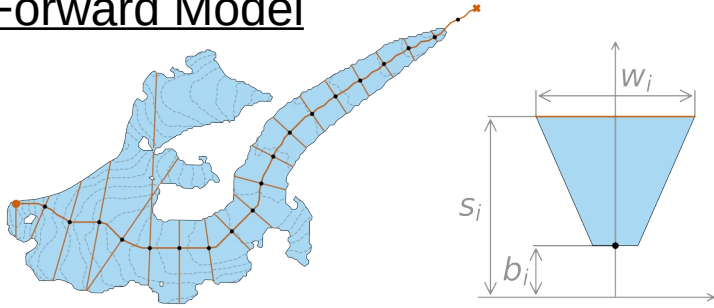
³School of Geosciences, University of Edinburgh, Edinburgh, United Kingdom

Introduction

- Started my PhD in September 2021,
in the PROVIDE Project www.provide-h2020.eu
as part of the Open Global Glacier Model (OGGM) Team
- **Goal:** Project mountain glacier contribution to freshwater availability for different scenarios.
- **Problems** (on a global scale):
 - Better projections require a data informed and consistent initial state!
 - Until recently one outline (with DEM) available \implies equilibrium assumption
 - But more data is getting available (geodetic mass-balance, surface ice velocity)
 - How to **incorporate** this **new data** to **initialize a transient glacier model**?



Forward Model



Cost Function

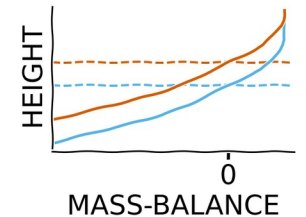
$$\mathcal{J}(\zeta) = \lambda_{dV} \frac{1}{\sigma_{dV}^2} (dV_m(\zeta) - dV_o)^2 + \lambda_s \frac{1}{n\sigma_s^2} \sum_{i=1}^n (s_{m,i}(\zeta) - s_{o,i})^2 + \frac{\lambda_b}{n} \sum_{i=1}^n \left(\frac{\partial b}{\partial x} \right)^2$$

Update ζ using
Automatic Differentiation
of PyTorch

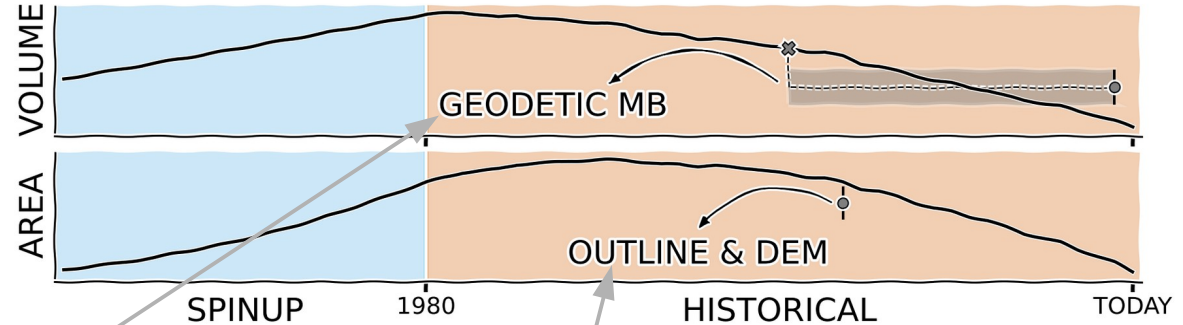
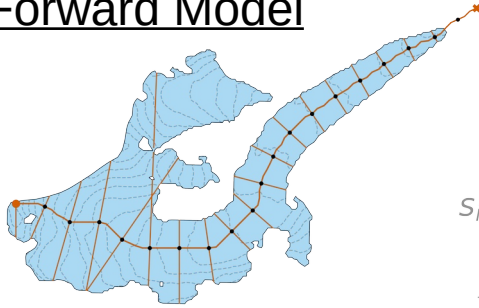
Glacier bed
heights \mathbf{b}_i

&

MB-SPINUP
height shift



Forward Model



Cost Function

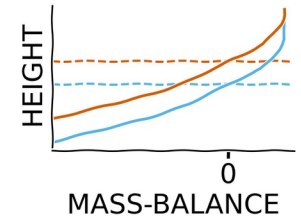
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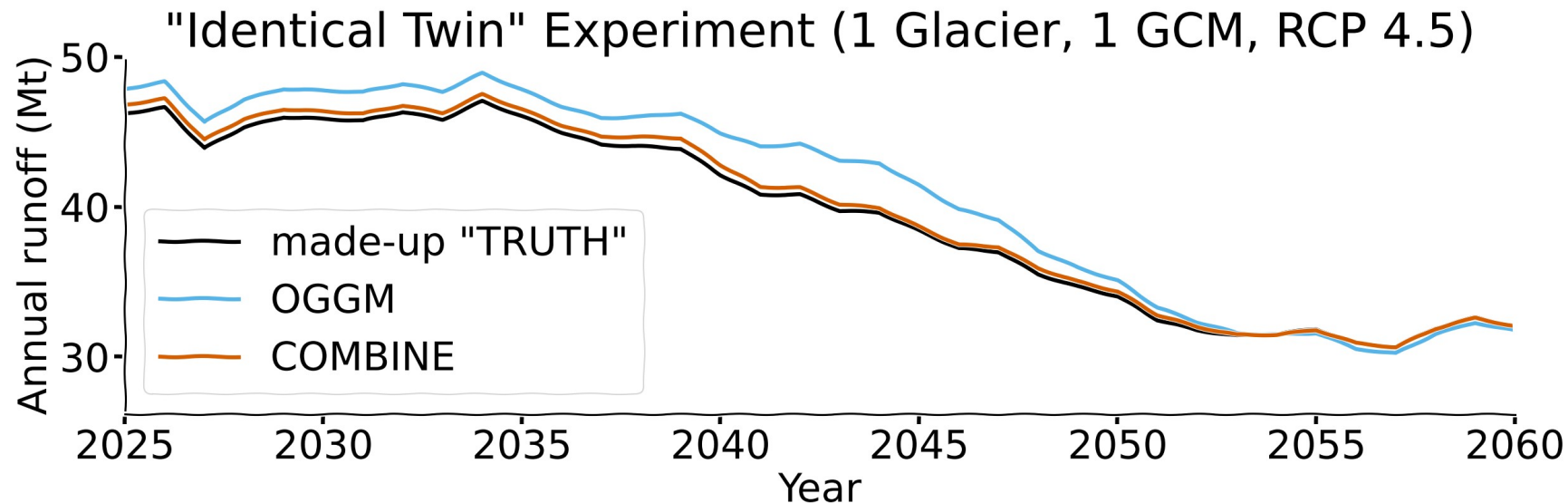
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height shift



(preliminary, unpublished) Results & Outlook



Upcoming Challenges

- Regularization methods
- Deal with model uncertainty
- Calibrate free model parameters
- Use surface ice velocity observations
- Include MB-model as control

Thank you for your attention!

Any questions or want to get in touch:

patrick.schmitt@uibk.ac.at

GitHub: github.com/OGGM/COMBINE

References:

- Horizon 2020 Project PROVIDE www.provide-h2020.eu
- Open Global Glacier Model OGGM oggm.org
- Paszke, A., and Coauthors, 2019: Pytorch: An imperative style, high-performance deep learning library. Advances in Neural Information Processing Systems 32, Curran Associates, <http://papers.neurips.cc/paper/9015-pytorch-an-imperative-style-high-performance-deep-learning-library.pdf>

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